- Title: Tree height and leaf drought tolerance traits shape growth responses across droughts in a temperate
- 2 broadleaf forest

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22 Summary

- As climate change drives increased drought in many forested regions, mechanistic understanding of
 the factors conferring drought tolerance in trees is increasingly important. The dendrochronological
 record provides a window through which we can understand how tree size and traits shape growth
 responses to droughts.
- We analyzed tree-ring records for twelve species in a broadleaf deciduous forest in Virginia (USA) to test hypotheses on how tree height, microenvironment characteristics, and species' traits shaped drought responses across the three strongest regional droughts over a 60-year period.
- Drought tolerance (resistance, recovery, and resilience) decreased with tree height, which was strongly correlated with exposure to higher evaporative demand and solar radiation. The potentially greater rooting volume of larger trees did not confer a resistance advantage, but marginally increased recovery and resilience, in sites with low topographic wetness index. Drought tolerance was greater among species whose leaves experienced less shrinkage upon desiccation and lost turgor (wilted) at more negative water potentials.
- The tree-ring record reveals that tree height and leaf drought tolerance traits influenced growth responses during and after significant droughts in the meteorological record. As climate change-induced droughts intensify, tall trees with drought-sensitive leaves will be most vulnerable to immediate and longer-term growth reductions.
- Key words: annual growth; crown exposure; drought; Forest Global Earth Observatory (ForestGEO); leaf drought tolerance traits; temperate broadleaf deciduous forest; tree height; tree-ring

Introduction

- Forests play a critical global role in climate regulation (Bonan, 2008), yet there remains enormous uncertainty as to how the forest-dominated terrestrial carbon sink will respond to climate change (Friedlingstein et al., 2006). An important aspect of this uncertainty lies with physiological responses of trees to drought (Kennedy et al., 2019). In many forested regions around the world, the risk of severe drought is increasing (Trenberth et al., 2014; Dai et al., 2018), often despite increasing precipitation 47 (Intergovernmental Panel on Climate Change, 2015; Cook et al., 2015). Droughts, intensified by climate change, have been affecting forests worldwide and are expected to continue as one of the most important drivers of forest change in the future (Allen et al., 2010, 2015; McDowell et al., 2020). Understanding forest responses to drought requires elucidation of how tree size, microenvironment, and species' traits jointly influence individual-level drought tolerance, defined here as a tree's ability to maintain growth during 52 drought (resistance) and to recover to its pre-drought growth rate (resilience) (Lloret et al., 2011). Because 53 the resistance and resilience of growth to drought are linked to trees' probability of surviving drought (DeSoto et al., 2020; Liu et al., 2019), understanding growth responses can also help elucidate which trees are most vulnerable to drought-induced mortality. However, it has proven difficult to resolve the many factors affecting tree growth during drought and the extent to which their influence is consistent across droughts. This is because available forest census data only rarely captures extreme drought, whereas tree-ring records capture multiple droughts but typically focus on only the largest individuals of one or a few species. Many studies have shown that within and across species, large trees tend to be more affected by drought. Greater growth reductions (i.e., lower drought resistance) in larger trees were first shown on a global scale by Bennett et al. (2015), and subsequent studies have reinforced this finding (e.g., Gillerot et al., 2020). 63
- Although lower recovery and resilience of larger trees have also been observed (Hacket-Pain et al., 2016;
- Gillerot et al., 2020), in general we have much more limited understanding of how and why these scale with tree size.
- physiological model of Trugman et al. (2018) predicts that recovery and resilience would not necessarily be
- lower, even in trees that will ultimately die. Moreover, it has yet to be resolved which of several potential underlying mechanisms most strongly shape
- these trends in drought response. First, tree height itself may be a primary driver. Taller trees face the biophysical challenge of lifting water greater distances against the effects of gravity and friction (McDowell et al., 2011; McDowell and Allen, 2015; Ryan et al., 2006; Couvreur et al., 2018). Vertical gradients in stem 72 and leaf traits-including smaller and thicker leaves (higher leaf mass per area, LMA), greater resistance to 73 hydraulic dysfunction (i.e., more negative water potential at 50% loss of hydraulic conductivity, more negative P50), and lower hydraulic conductivity at greater heights (Couvreur et al., 2018; Koike et al.,
- 2001; McDowell et al., 2011)—enable trees to become tall (Couvreur et al., 2018). Greater stem capacitance 76 (i.e., water storage capacity) of larger trees may also confer resistance to transient droughts (Phillips et al.,
- 2003; Scholz et al., 2011). Taller trees have wider conduits in the basal portions of taller trees, both within
- and across species (Olson et al., 2018; Liu et al., 2019) and throughout the conductive systems of 79
- angiosperms (Zach et al., 2010; Olson et al., 2014, 2018), which help maintain constant the resistance that
- would otherwise increase as trees grow taller. Wider xylem conduits plausibly make large trees more
- vulnerable to embolism during drought (Olson et al., 2018), and traits conducive to efficient water

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transport may also lead to poor ability to recover from or re-route water around embolisms (Roskilly et al.,
    2019). (here would be a good place to comment on resilience. "What may help the tree in
    competition, therefore, can also be a deteriment to drought resilience.")
    Larger trees may also have lower drought tolerance because of microenvironmental and ecological factors.
    Their crowns tend to occupy more exposed canopy positions, which are associated with higher evaporative
87
    demand (Kunert et al., 2017). Subcanopy trees tend to fare better specifically due to the benefits of a
    buffered environment (Pretzsch et al., 2018). Counteracting the liabilities associated with tall height, large
    trees tend to have larger root systems (Enquist and Niklas, 2002; Hui et al., 2014), potentially mitigating
    some of the biophysical challenges they face by allowing greater access to water. Larger root systems-if
    they grant access to deeper water sources—would be particularly advantageous in drier microenvironments
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    (e.g., hilltops, as compared to valleys and streambeds) during drought. Finally, tree size-related responses
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    to drought can be modified by species' traits and their distribution across size classes (Meakem et al., 2018;
    Liu et al., 2019). Understanding the mechanisms driving the greater relative growth reductions of larger
    trees during drought requires sorting out the interactive effects of height and associated exposure, root
    water access, and species' traits.
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    Debates have also arisen regarding the traits influencing tree growth responses to drought. Studies within
    temperate broadleaf forests have observed ring-porous species showing higher drought tolerance than
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    diffuse-porous species (Friedrichs et al., 2009; Elliott et al., 2015; Kannenberg et al., 2019), but this
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    distinction does not always hold (Martin-Benito and Pederson, 2015), nor does it hold in the global context
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    (Wheeler et al., 2007; Olson et al., 2020), and it does not resolve differences among the many species
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    within each category. Commonly-measured traits including wood density and leaf mass per area (LMA)
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    have been linked to drought responses within some temperate deciduous forests (Abrams, 1990; Guerfel
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    et al., 2009; Hoffmann et al., 2011; Martin-Benito and Pederson, 2015) and across forests worldwide
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    (Greenwood et al., 2017). However, in other cases these traits could not explain drought tolerance (e.g., in
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    a tropical rainforest; Maréchaux et al., 2019), or the direction of response was not always consistent. For
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    instance, higher wood density has been associated with greater drought resistance at a global scale
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    (Greenwood et al., 2017), but correlated negatively with tree performance during drought in a broadleaf
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    deciduous forest in the southeastern United States (Hoffmann et al., 2011). Thus, the perceived influence
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    of these traits on drought resistance may actually reflect indirect correlations with other traits that more
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    directly drive drought responses (Hoffmann et al., 2011).
    In contrast, hydraulic traits have direct physiological linkages to tree growth and mortality responses to
    drought. For instance, water potentials at which percent the loss of conductivity surpasses a certain
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    threshold (e.g., P50 and P88, representing 50 and 88% loss of conductivity, respectively) and hydraulic
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    safety margin (i.e., difference between typical minimum water potentials and P50 or P88) correlate with
    drought performance across global forests (Anderegg et al., 2016). However, these are time-consuming to
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    measure and therefore infeasible for predicting or modeling drought responses in highly diverse forests
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    (e.q., in the tropics). More easily-measurable leaf drought tolerance traits that have direct linkage to plant
    hydraulic function can explain variation in plant distribution and function (Medeiros et al., 2019). These
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    include leaf area shrinkage upon desiccation (PLA_{dry}; Scoffoni et al., 2014) and the leaf water potential at
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    turgor loss point (\pi_{tlp}), i.e., the water potential at which leaf wilting occurs (Bartlett et al., 2016a; Zhu
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    et al., 2018). Both traits correlate with hydraulic vulnerability and drought tolerance as part of unified
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plant hydraulic systems (Scoffoni et al., 2014; Bartlett et al., 2016a; Zhu et al., 2018; Farrell et al., 2017).

The abilities of both PLA_{dry} and π_{tlp} to explain the drought tolerance of tree growth remains untested. 125 Here, we examine how tree height, microenvironment characteristics, and species' traits collectively shape 126 three metrics of drought tolerance: (1) resistance, defined as the ratio of annual stem growth in a drought year to that which would be expected in the absence of drought based on previous years' growth; (2) 128 recovery, defined the ratio of post-drought growth to growth during the drought year; and (3) resilience, 129 defined as the ratio of post-drought to pre-drought growth (Lloret et al., 2011). We test a series of hypotheses and associated specific predictions (Table 1) based on the combination of tree-ring records from 131 the three strongest droughts over a 60-year period (1950 - 2009), species trait measurements, and census 132 and microenvironmental data from a large forest dynamics plot in Virginia, USA. First, we focus on how tree size, alone and in its interaction with microenvironmental gradients, influences drought tolerance. We 134 examine the contemporary relationship between tree height and microenvironment, including growing 135 season meteorological conditions and crown exposure. We then test whether, consistent with most forests 136 globally, larger-diameter, taller trees tend to have lower drought tolerance in this forest, which is in a 137 region (eastern North America) represented by only two studies in the global review of (Bennett et al., 138 2015). We also test for an influence of potential access to available soil water, which should be greater for 139 larger trees in dry but not in perpetually wet microsites. Finally, we focus on the role of species' traits, 140 testing the hypothesis that species' traits-particularly leaf drought tolerance traits-predict drought 141 tolerance. We test predictions that drought tolerance is higher in ring-porous than semi-ring and diffuse-porous species and that it is correlated with wood density-either positively (Greenwood et al., 2017) 143 or negatively (Hoffmann et al., 2011) and positively correlated with LMA. We further test predictions that 144 species with low PLA_{dry} and those whose leaves lose turgor at lower water potentials (more negative π_{tlp}) 145 have higher tolerance. 146

147 Materials and Methods

148 Study site and microclimate

Research was conducted at the 25.6-ha ForestGEO (Forest Global Earth Observatory) study plot at the
Smithsonian Conservation Biology Institute (SCBI) in Virginia, USA (38°53'36.6"N, 78°08'43.4"W; Fig.
S1) (Bourg et al., 2013; Anderson-Teixeira et al., 2015a). SCBI is located in the central Appalachian
Mountains near the northern boundary of Shenandoah National Park. Elevations range from 273 to 338 m
above sea level with a topographic relief of 65m (Bourg et al., 2013). Climate is humid temperate, with
mean annual temperature of 12.7°C and precipitation of 1005 mm yr⁻¹ during our study period (1960-2009;
source: CRU TS v.4.01; Harris et al., 2014). Dominant tree taxa within this secondary forest include
Liriodendron tulipifera, oaks (Quercus spp.), and hickories (Carya spp.; Table 2).

157 Identifying drought years

We identified the three largest droughts within the time period 1950-2009, defining drought (Slette et al., 2019) as events with anomalously dry peak growing season climatic conditions. Specifically, we used the metric of Palmer Drought Severity Index (PDSI) during May-August (MJJA; Table S1), which were identified by Helcoski et al. (2019) as the months of the current year to which annual tree growth was most sensitive at this site. PDSI divisional data for Northern Virginia were obtained from NOAA (https://www7.ncdc.noaa.gov/CDO/CDODivisionalSelect.jsp) in December 2017. Based on this, we identified the three strongest droughts during the study period (Figs. 1, S1; Table S1).

The droughts differed in intensity and antecedent moisture conditions (Fig. S1, Table S1). The 1966 drought was preceded by two years of moderate drought during the growing season and severe to extreme drought starting the previous fall. In August 1966, PDSI reached its lowest monthly value (-4.82) of the three droughts. The 1977 drought was the least intense throughout the growing season, and it was preceded by 2.5 years of near-normal conditions, making it the mildest of the three droughts. The 1999 drought was preceded by wetter than average conditions until the previous June, but PDSI plummeted below -3.0 in October 1998 and remained below this threshold through August 1999. Following all three droughts, PDSI rebounded to near-normal conditions in September or October (Fig. S1).

Data collection and preparation

Within or just outside the ForestGEO plot, we collected data on a suite of variables including tree heights, microenvironment characteristics, and species traits (Table 3). The SCBI ForestGEO plot was censused in 2008, 2013, and 2018 following standard ForestGEO protocols, whereby all free-standing woody stems \geq 1cm diameter at breast height (DBH) were mapped, tagged, measured at DBH, and identified to species (Condit, 1998). From these census data, we used measurements of DBH from 2008 to calculate historical DBH and data for all stems \geq 10cm to analyze functional trait composition relative to tree height (all analyses described below). Census data are available through the ForestGEO data portal (www.forestgeo.si.edu).

We analyzed tree-ring data (xylem growth increment) from 571 trees representing the twelve dominant 182 species (Table 2; Fig. S2). Selected species were those with the greatest contributions to woody 183 aboveground net primary productivity $(ANPP_{stem})$ and together comprised 97% of study plot $ANPP_{stem}$ 184 between 2008 and 2013 (Helcoski et al., 2019). Cores (one per tree) were collected within the ForestGEO 185 plot at breast height (1.3m) in 2010-2011 or 2016-2017. In 2010-2011, cores were collected from randomly 186 selected live trees of each species that had at least 30 individuals ≥ 10 cm DBH (Bourg et al., 2013). 187 Annual tree mortality censuses were initiated in 2014 (Gonzalez-Akre et al., 2016), and in 2016-2017, cores were collected from all trees found to have died since the previous year's census. We note that drought was 189 probably not a cause of mortality for these trees, as monthly May-Aug PDSI did not drop below -1.75 in 190 these years or the three years prior (2013-2017), and that trees cored dead displayed similar climate sensitivity to trees cored live (Helcoski et al., 2019). Lagged drought-induced mortality would be unlikely, 192 given that the trees analyzed here lived at least 17-18 years past the most recent major drought (1999), 193 whereas the meta-analysis of Trugman et al. (2018) indicates that >10-year lags in drought-attributed 194 mortality are rare. Cores were sanded, measured, and crossdated using standard procedures, as detailed in 195 (Helcoski et al., 2019). The resulting chronologies (Fig. 1a) were published in Zenodo (Gonzalez-Akre 196 et al., 2019). 197

For each cored tree, we combined tree-ring records and allometric equations of bark thickness to reconstruct DBH for the years 1950-2009. Prior *DBH* was estimated using the following equation:

$$DBH_{Y} = DBH_{2008} - 2 * \left[r_{bark,2008} - r_{bark,Y} + \sum_{year=Y}^{2008} r_{ring,Y} \right]$$

Here, Y denotes the year of interest, r_{ring} denotes ring width derived from cores, and r_{bark} denotes bark thickness, which was estimated from species-specific allometries based on the bark thickness data from the site (Table S2; Anderson-Teixeira et al., 2015b).

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Tree heights (H) were measured by several researchers for a variety of purposes between 2012 and 2019
    (n=1,518 trees). Methods included direct measurements using a collapsible measurement rod on small trees
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    (NEON, 2018) or a tape measure on recently fallen trees (this study); geometric calculations using
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    clinometer and tape measure (Stovall et al., 2018a) or digital rangefinders (Anderson-Teixeira et al., 2015b;
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    NEON, 2018); and ground-based LiDAR (Stovall et al., 2018b). Rangefinders used either the tangent
    method (Impulse 200LR, TruPulse 360R) or the sine method (Nikon ForestryPro) for calculating heights.
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    Both methods are associated with some error (Larjavaara and Muller-Landau, 2013), but in this instance
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    there was no clear advantage of one or the other. Measurements from the National Ecological Observatory
    Network (NEON) were collected near the ForestGEO plot following standard NEON protocol, whereby
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    vegetation of short stature was measured with a collapsible measurement rod, and taller trees with a
212
    rangefinder (NEON, 2018). Species-specific height allometries were developed using log-log regression
    (\ln[H] \sim \ln[DBH]; \text{ Table S3}). For species with insufficient height data to create reliable species-specific
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    allometries (n=2, JUNI and FRAM), heights were calculated from an equation developed by combining the
215
    height measurements across all species. We then used these allometries to estimate H for each drought
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    year, Y, based on reconstructed DBH_Y (Fig. S3).
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    To characterize how environmental conditions vary with height, data were obtained from the NEON tower
218
    located <1km from the study area via the neonUtilities package (Lunch et al., 2020). We used wind speed,
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    relative humidity, and air temperature data, all measured over a vertical profile spanning heights from 7.2
    m to above the top of the tree canopy (31.0 or 51.8m, depending on censor), for the years 2016-2018
221
    (NEON, 2018). After filtering for missing and outlier values, we determined the daily minima and maxima.
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    which we then aggregated at the monthly scale.
223
    Crown position—a categorical variable classifying trees based on exposure to sunlight—was recorded for all
224
    cored trees that remained standing during the growing season of 2018 following the protocol of Jennings
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    et al. (1999). Trees were classified as follows: dominant trees were defined as those with crowns above the
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    general level of the canopy, co-dominant trees as those with crowns within the canopy; intermediate
227
    trees as those with crowns below the canopy level, but illuminated from above; and suppressed as those
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    below the canopy and receiving minimal direct illumination from above.
229
    Topographic wetness index (TWI), used here as a metric of long-term mean moisture availability, was
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    calculated using the dynatopmodel package in R (Fig. S2) (Metcalfe et al., 2018). Originally developed by
231
    Beven and Kirkby (1979), TWI was part of a hydrological run-off model and has since been used for a
232
    number of purposes in hydrology and ecology (Sørensen et al., 2006). TWI calculation depends on an input
    of a digital elevation model (DEM; ~3.7 m resolution from the elevatr package (Hollister, 2018)), and from
234
    this yields a quantitative assessment defined by how "wet" an area is, based on areas where run-off is more
235
    likely. From our observations in the plot, TWI performed better at categorizing wet areas than the
    Euclidean distance from the stream.
237
    Species' trait data were collected in August 2018 (Tables 2-3; Fig. S4). We sampled small, sun-exposed
238
    branches up to eight meters above the ground from three individuals of each species in and around the
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    ForestGEO plot. Sampled branches were re-cut under water at least two nodes above the original cut and
    re-hydrated overnight in covered buckets under opaque plastic bags before measurements were taken.
241
    Rehydrated leaves taken towards the apical end of the branch (n=3 per individual: small, medium, and
242
    large) were scanned, weighed, dried at 60° C for ≥ 48 hours, and then re-scanned and weighed. Leaf area
243
    was calculated from scanned images using the LeafArea R package (Katabuchi, 2019). LMA was
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calculated as the ratio of leaf dry mass to fresh area. PLA_{dry} was calculated as the percent loss of area
245
    between fresh and dry leaves. Wood density was calculated for ~1cm diameter stem samples (bark and pith
    removed) as the ratio of dry weight to fresh volume, which was estimated using Archimedes' displacement.
247
    We used the rapid determination method of Bartlett et al. (2012) to estimate osmotic potential at turgor
248
    loss point (\pi_{tlp}). Briefly, two 4 mm diameter leaf discs were cut from each leaf, tightly wrapped in foil,
    submerged in liquid nitrogen, perforated 10-15 times with a dissection needle, and then measured using a
250
    vapor pressure osmometer (VAPRO 5520, Wescor, Logan, UT, USA). Osmotic potential (\pi_{osm}) given by
251
    the osmometer was used to estimate (\pi_{tlp}) using the equation \pi_{tlp} = 0.832\pi_{osm}^{-0.631} (Bartlett et al., 2012).
252
    Statistical Analysis
253
254
    For each drought year, we calculated metrics of drought resistance (Rt), recovery (Rc), and resilience (Rs),
    following Lloret et al. (2011) (Table 3). We defined Rt as the ratio of basal area increment (BAI; i.e.,
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    change in cross-sectional area) during the drought year to the mean BAI over the five years preceding the
256
    drought. Rc was defined as the ratio of mean BAI over the five years following the drought to BAI of the
257
    drought year. Rs was defined as the ratio of mean BAI over the five years following the drought to that of
258
    the five years preceding the drought. Thus, for all metrics, values <1 and >1 indicate growth reductions
259
    and increases, respectively.
260
    Because these metric could potentially be biased by directional pre-drought growth trends, we also tried an
261
    intervention time series analysis (ARIMA, Hyndman et al., 2020) that predicted mean drought-year growth
262
    based on trends over the preceding ten years and used this value in place of the five-year mean in
263
    calculations of resistance (Rt_{ARIMA} = observed BAI/ predicted BAI). Rt and Rt_{ARIMA} were strongly
264
    correlated (Fig. S5), and showed similar responses to the independent variables of interest (cf. Tables
265
    S4-55, S8-S9). Visual review of the individual tree-ring sequences with the largest discrepancies between
    these metrics revealed that Rt was less prone to unreasonable estimates than Rt_{ARIMA}. We therefore
267
    determined that use of 5-year means, as described above, were more appropriate metrics than those based
268
    on ARIMA projections.
269
    Analyses focused on testing the predictions presented in Table 1 with Rt (or Rt_{ARIMA}), Rc, or Rs as the
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    response variable. Models were run for all drought years combined and for each drought year individually.
271
    The general statistical model for hypothesis testing was a mixed effects model, implemented in the lme4
    package in R (Bates et al., 2019). In the multi-year model, we included a random effect of tree nested
273
    within species and a fixed effect of drought year to represent the combined effects of differences in drought
274
    characteristics. Individual year models included a random effect of species. All models included fixed
275
    effects of independent variables of interest (Tables 1,3) as specified below. All variables across all best
276
    models had variance inflation factors between 1 and 1.045. We used Akaike information criterion with
277
    correction for small sample sizes (AICc; see Brewer et al., 2016) to assess model selection, and
278
    conditional/marginal R-squared to assess model fit as implemented in the AICcmodavg package in R
279
    (Mazerolle and portions of code contributed by Dan Linden., 2019). Individual model terms were
280
    considered significant when their addition to a model improved fit at \Delta AICc \geq 2.0, where \Delta AICc is the
281
    difference in AICc between models with and without the trait.
282
    To avoid over-fitting models with five species traits (Table 3) across only 12 species, we did not include all
    traits as fixed effects in a single linear mixed model, but rather conducted individual tests of each species
284
    trait to determine the relative importance and appropriateness for inclusion in the main model. These tests
285
    followed the model structure specified above, then added ln[H] and ln[TWI] to create a base model against
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addition to the base model significantly improved fit for at least one metric of drought tolerance (Rt, Rc, or Rs; Tables S4,S6-S7). While we tested xylem porosity as a predictor (Table 1), we did not consider it 289 appropriate for inclusion in the main model because of highly uneven distribution of species across 290 categories (Table 2) and opposite drought responses of the only two diffuse-porous species (detailed below). We then determined the top full models for predicting each dependent variable. To do so, we compared models with all possible combinations of candidate variables, including ln[H]*ln[TWI] and species traits as 293 specified above. We identified the full set of models within $\triangle AICc=2$ of the best model (that with lowest 294 AICc). When a variable appeared in all of these models and the sign of the coefficient was consistent across models, we viewed this as support for the acceptance/rejection of the associated prediction (Table 1). If the variable appeared in some but not all of these models, and its sign was consistent across models, we 297 considered this partial support/rejection. All analysis beyond basic data collection was performed using R version 3.6.2 (R Core Team, 2019). Other R-packages used in analyses are listed in the Supplementary Information (Appendix S1). All data, code, 300 and results are available through the SCBI-ForestGEO organization on GitHub 301 (https://github.com/SCBI-ForestGEO: SCBI-ForestGEO-Data and 302 McGregor climate-sensitivity-variation repositories), with static versions corresponding to data and 303 analyses presented here archived in Zenodo (DOIs: 10.5281/zenodo.3604993 and [TBD], respectively.

which we tested traits. Trait variables were considered appropriate for inclusion in the main model if their

Results 305

Tree height and microenvironment

In the years for which we have vertical profiles in climate data (2016-2018), taller trees—or those in 307 dominant crown positions—were generally exposed to higher evaporative demand during the peak growing 308 season months (May-August; Fig. 2). Specifically, maximum daily wind speeds were significantly higher above the top of the canopy (40-50m) than within and below (10-30m) (Fig. 2a). Relative humidity was 310 also somewhat lower during June-August, ranging from ~50-80\% above the canopy and ~60-90\% in the 311 understory (Fig. 2b). Air temperature did not vary consistently across the vertical profile (Fig. 2c). 312 Crown position varied as expected with height (dominant > co-dominant > intermediate > suppressed), but with substantial variation (Fig. 2d). There were significant differences in height across all crown 314 position classes (Fig. 2d). A comparison test between height and crown position data from the most recent 315 ForestGEO census (2018) revealed a correlation of 0.73. 316 Community-level drought responses

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At the community level, cored trees showed substantial growth reductions in all three droughts, with a 318 mean Rt of 0.86 in 1966 and 1999, and 0.84 in 1977 (Fig. 1b). Across the entire study period (1950-2009), 319 the focal drought years were the three years with the largest fraction of trees exhibiting $Rt \leq 0.7$. Specifically, in each drought, roughly 30% of the cored trees had growth reductions of $\geq 30\%$ ($Rt \leq 0.7$): 321 29% in 1966, 32% in 1977, and 27% in 1999. However, some individuals exhibited increased growth, i.e., 322 Rt > 1.0: 26% of trees in 1966, 22% in 1977, and 26% in 1999. Recovery was generally strong and complete, with Rc averaging 1.55 in 1966, 1.42 in 1977, and 1.34 in 1999 (Fig. S6) and Rs averaging 1.28 324 in 1966, 1.19 in 1977, and 1.12 in 1999 (Fig. 1c).

In the context of the multivariate model, Rt did not vary across drought years. That is, drought year as a variable did not appear in any of the top models -i.e., models that were statistically indistinguishable $(\Delta AICc < 2)$ from the best model (see footnotes on Tables S8-S9). In contrast, both Rc and Rs varied 328 across years, being highest in all of the top models (Tables S10-S11), in which both were highest in 1966, 329 intermediate in 1977, and lowest in 1999. Tree height, microenvironment, and drought tolerance 331 Taller trees (based on H in the drought year) showed stronger growth reductions during drought (i.e., 332 lower Rt) and less resilience following drought (i.e., lower Rc and Rs; Table 1; Fig. 4). Specifically, for Rt, 333 ln[H] appeared, with negative coefficient, in the best model ($\Delta AICc=0$) and all top models when evaluating the three drought years together (Tables S8-S9). The same held true for 1966 individually, but 335 there was no significant effect of ln[H] for 1977 or 1999 individually. For Rc, ln[H] appeared, with negative 336 coefficient, in the best model without a ln[H] * ln[TWI] interaction, for the three drought years together and for 1977, but not for 1966 or 1999. For Rs, again considering the best models without a 338 ln[H] * ln[TWI] interaction, there was a negative effect of ln[H] for the three drought years together and 339 for 1966 and 1977, and a non-significant negative trend in 1999. Trees in drier microsites showed greater growth declines during drought; i.e., Rt had a significantly negative 341 response to ln[TWI] across all drought years combined, and in 1977 and 1999 individually (Fig. 4, Table 342 S8-S9). The ln[H] * ln[TWI] interaction was never significant, and had a positive sign in any top Rt models in which it appeared (Tables 1, S8-S9), rejecting the hypothesis that smaller trees (presumably with 344 smaller rooting volume) are more susceptible to drought in microenvironments with a deeper water table. 345 In contrast, ln[TWI] did not appear in any of the best models for Rc or Rs (combined of for individual 346 years), except in interaction with ln[H] (Fig. 4, Tables S10-S11). Negative ln[H] * ln[TWI] interactions 347 appeared in the best models for both Rc and Rs for all years combined, as well as in one individual year 348 for each (1966 for Rc, 1977 for Rs). This implies that small trees had greater recovery and resilience in 349 wetter microhabitats, whereas large trees had greater recovery and resilience in dry microhabitats. Species' traits and drought tolerance 351 Species, as a factor in ANOVA, had significant influence (p< 0.05) on all traits (wood density, LMA, 352 PLA_{dry} , and π_{tlp}), with more significant pairwise differences for wood density and PLA_{dry} than for LMA353 and π_{tlp} (Table 2, Fig. S4 as characterized by the agricolae package de Mendiburu (2020)). Drought 354 tolerance also varied across species, overall and in each drought year (Figs. 3, S7). Species with overall 355 lowest and highest Rt and Rs were, respectively, Liriodendron tulipifera (mean Rt = 0.66, mean Rs = 0.66) 356 1.04) and Fagus grandifolia (mean Rt = 0.99; mean Rs = 1.65). These two species—notably the only two diffuse-porous species in our study-differed significantly from one another in Rt and Rs in each drought 358 year (Fig. 3). Wood density, LMA, and xylem porosity were all poor predictors of drought tolerance (Tables 1,S4-S5). Wood density and LMA were never significantly associated with Rt, Rc, or Rs in the single-variable tests 361 and were therefore excluded from the full models. Xylem porosity had no significant influence on Rt or Rs362 in models for all droughts combined (Tables S4, S7). In contrast, Rc was significantly higher in diffuse- and

Drought resistance and resilience, but not recovery, were negatively correlated with PLA_{dry} and π_{tlp} (Fig. 4; Tables 1, S4-S11). For Rt, PLA_{dry} had a significant influence, with negative coefficient, in top models

semi-ring porous species than in ring-porous species (Table S6, Fig. 3).

364

included in some of the top models for 1999 (Tables S8-S9). π_{tlp} was included with a negative coefficient in the best model for both all droughts combined and for the 1977 drought individually (Fig. 4; Table S8), although its influence was not significant at $\Delta \text{AICc} < 2$. It was also included in some of the top models for 1999 (Tables S8-S9). Recovery was not significantly correlated with either PLA_{dry} or π_{tlp} . There was only one best Rc model containing one of these terms (π_{tlp} in 1977 drought), but in no instance was one of these terms included in all top models (i.e., at $\Delta \text{AICc} < 2$).

for the three droughts combined and for the 1966 drought individually (Fig. 4; Tables S8-S9). It was also

For Rs, PLA_{dry} was in the best models for the three droughts combined and for the 1966 drought individually, and some of the top models for 1977 and 1999 (Fig. 4; Table S11); however, its effects were not significant atat $\Delta AICc < 2$. π_{tlp} was in the best models for the three droughts combined and for 1966 and 1999 individually, and in one of the top models for 1977 (Fig. 4; Table S11). It's effects were significant at $\Delta AICc < 2$ for 1999 only.

380 Discussion

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Tree height, microenvironment, and leaf drought tolerance traits shaped tree growth responses across three 381 droughts at our study site (Table 1, Fig. 4). The greater susceptibility of larger trees to drought, similar to forests worldwide (Bennett et al., 2015), was driven primarily by their height (Stovall et al., 2019). Taller 383 height was likely a liability in itself, and was also associated with greater exposure to conditions that would 384 promote water loss and heat damage during drought (Fig. 2). There was no evidence that greater availability of, or access to, soil water availability increased drought resistance; in contrast, trees in wetter 386 topographic positions had lower Rt (Zuleta et al., 2017; Stovall et al., 2019), and the larger potential 387 rooting volume of large trees provided no advantage in the drier microenvironments. The negative effect of 388 height on Rt held after accounting for species' traits, which is consistent with recent work finding height 389 had a stronger influence on mortality risk than forest type during drought (Stovall et al., 2020). Drought 390 tolerance was not consistently linked to species' LMA, wood density, or xylem type (ring- vs. diffuse porous), but was negatively correlated with leaf drought tolerance traits (PLA_{dry}, π_{tlp}) . This is the first 392 study to our knowledge linking PLA_{dry} and π_{tlp} to growth reduction during drought. The directions of 393 these responses were consistent across droughts (Table S8), supporting the premise that they were driven 394 by fundamental physiological mechanisms. However, the strengths of each predictor varied across droughts 395 (Fig. 4; Tables S8-S9), indicating that drought characteristics interact with tree size, microenvironment, 396 and traits to shape which individuals are most affected. These findings advance our knowledge of the factors that make trees vulnerable to stem growth declines during drought and, by extension, likely make 398 them more vulnerable to mortality (Sapes et al., 2019). 399 The droughts considered here were of a magnitude that has occurred with an average frequency of 400 approximately once every 10-15 years (Fig. 1a, Helcoski et al., 2019) and had substantial but not 401 devastating impacts on tree growth (Figs. 1b). These droughts were classified as severe (PDSI < -3.0; 402 1977) or extreme (PDSI < -4.0; 1966, 1999) at our site and have been linked to tree mortality in the 403 eastern United States (Druckenbrod et al., 2019). However, extreme, multiannual droughts such as the so-called "megadroughts" of this type that have triggered massive tree die-off in other regions (e.g., Allen 405 et al., 2010; Stovall et al., 2019) have not occurred in the Eastern United States within the past several

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decades (Clark et al., 2016). Of the droughts considered here, the 1966 drought, which was preceded by
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    two years of dry conditions (Fig. S1), severely stressed a larger portion of trees (Fig. 1b). The tendency
    for large trees to have lowest resistance was most pronounced in this drought, consistent with other
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    findings that this physiological response increases with drought severity (Bennett et al., 2015; Stovall et al.,
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    2019). Across all three droughts, the majority of trees experienced reduced growth, but a substantial
    portion (e.g., short understory trees, species with drought resistant traits; Fig. 4) had increased growth
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    (Fig. 1b), consistent with prior observations that smaller trees can exhibit increased growth rates during
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    drought (Bennett et al., 2015). It is likely because of the moderate impact of these droughts, along with
    other factors influencing tree growth (e.g., stand dynamics), that our best models characterize only a
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    modest amount of variation in Rt: 11-12% for all droughts combined, and 18-25% for each individual
416
    drought (Table S8).
    Consistent with studies in other forests worldwide (Bennett et al., 2015), taller trees in this forest exhibited
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    lower drought resistance—and also recovery and resilience. Mechanistically, this is consistent with, and
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    reinforces, previous findings that it impossible for trees to efficiently transport water to great heights and
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    simultaneously maintain strong resistance and resilience to drought-induced embolism (Olson et al., 2018;
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    Couvreur et al., 2018; Roskilly et al., 2019). Taller trees also face dramatically different microenvironments
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    (Fig. 2). They are exposed to higher wind speeds and lower humidity (Fig. 2a-b), resulting in higher
423
    evaporative demand. Unlike other temperate forests where modestly cooler understory conditions have
424
    been documented (Zellweger et al., 2019), particularly under drier conditions (Davis et al., 2019), we
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    observed no significant variation in air temperatures across the vertical profile (Fig. 2c). More critically for
    tree physiology, leaf temperatures can become significantly elevated over air temperature under conditions
427
    of high solar radiation and low stomatal conductance (Campbell and Norman, 1998; Rey-Sánchez et al.,
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    2016). Under drought, when air temperatures tend to be warmer, direct solar radiation tends to be higher
    (because of less cloud cover), and less water is available for evaporative cooling of the leaves, trees with
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    sun-exposed crowns may not be able to simultaneously maintain leaf temperatures below damaging
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    extremes and avoid drought-induced embolism. Indeed, previous studies have shown lower drought
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    resistance in more exposed trees (Liu and Muller, 1993; Suarez et al., 2004; Scharnweber et al., 2019).
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    Unfortunately, collinearity between height and crown exposure in this study (Fig. 2d) makes it impossible
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    to confidently partition causality. Additional research comparing drought responses of early successional
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    and mature forest stands, along with short and tall isolated trees, would be valuable for more clearly
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    disentangling the roles of tree height and crown exposure.
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    Belowground, taller trees would tend to have larger root systems (Enquist and Niklas, 2002; Hui et al.,
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    2014), but this does not necessarily imply that they have greater access to or reliance on deep soil-water
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    resources that may be critical during drought. While tree size can correlate with the depth of water
    extraction (Brum et al., 2019), the linkage is not consistent. Shorter trees can vary broadly in the depth of
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    water uptake (Stahl et al., 2013), and larger trees may allocate more to abundant shallow roots that are
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    beneficial for taking up water from rainstorms (Meinzer et al., 1999). Moreover, reliance on deep soil-water
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    resources can actually prove a liability during severe and prolonged drought, as these can experience more
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    intense water scarcity relative to non-drought conditions (Chitra-Tarak et al., 2018). In any case, the
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    potentially greater access to water did not override the disadvantage conferred by height-and, in fact,
    greater moisture access in non-drought years (here, higher TWI) appears to make trees more sensitive to
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    drought (Zuleta et al., 2017; Stovall et al., 2019). This may be because moister habitats would tend to
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    support species and individuals with more mesophytic traits (Bartlett et al., 2016b; Mencuccini, 2003;
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Medeiros et al., 2019), potentially growing to greater heights [e.g., Detto et al. (2013), and these are then
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    more vulnerable when drought hits. The observed height-sensitivity of Rt, together with the lack of
    conferred advantage to large stature in drier topographic positions, agrees with the concept that
452
    physiological limitations to transpiration under drought shift from soil water availability to the
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    plant-atmosphere interface as forests age (Bretfeld et al., 2018), such that tall, dominant trees are the most
    sensitive in mature forests. Again, additional research comparing drought responses across forests with
455
    different tree heights and water availability would be valuable for disentangling the relative importance of
456
    above- and belowground mechanisms across trees of different size.
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    The development of tree-ring chronologies for the twelve most dominant tree species at our site (Helcoski
458
    et al., 2019; Bourg et al., 2013) gave us the sample size to compare historical drought responses across
459
    species (Fig. 3) and associated traits at a single site (see also Elliott et al., 2015). Our study reinforced
    current understanding (see Introduction) that wood density and LMA are not reliably linked to drought
461
    resistance (Table 1). Contrary to several previous studies in temperate deciduous forests (Friedrichs et al.,
462
    2009; Elliott et al., 2015; Kannenberg et al., 2019), we did not find an association between xylem porosity
463
    and drought resistance, as the two diffuse-porous species, Liriodendron tulipifera and Faqus grandifolia,
464
    were at opposite ends of the Rt spectrum (Fig. 3). While the low Rt of L. tulipifera is consistent with
465
    other studies (Elliott et al., 2015), the high Rt of F. grandifolia contrasts with studies identifying diffuse
    porous species in general (Elliott et al., 2015; Kannenberg et al., 2019), and the genus Faqus in particular
467
    (Friedrichs et al., 2009), as drought sensitive. There are two potential explanations for this discrepancy.
468
    First, other traits can and do override the influence of xylem porosity on drought resistance. Ring-porous
    species are restricted mainly to temperate deciduous forests, while highly drought-tolerant diffuse-porous
470
    species exist in other biomes (Wheeler et al., 2007). Fagus grandifolia had intermediate \pi_{tlp} and low
471
    PLA_{dry} (Fig. S4), which would have contributed to its drought resistance (Fig. 4; see discussion below).
    A second explanation of why F. grandifolia trees at this particular site had higher Rt is that the sampled
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    individuals, reflective of the population within the plot, are generally shorter and in less-dominant canopy
474
    positions compared to most other species (Fig. S4). The species, which is highly shade-tolerant, also has
    deep crowns (Anderson-Teixeira et al., 2015b), implying that a lower proportion of leaves would be affected
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    by harsher microclimatic conditions at the top of the canopy under drought (Fig. 2). Thus, the high Rt of
477
    the sampled F. grandifolia population can be explained by a combination of fairly drought-resistant leaf
478
    traits, shorter stature, and a buffered microenvironment.
479
    Concerted measurement of tree-rings and leaf drought tolerance traits of emerging importance (Scoffoni
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    et al., 2014; Bartlett et al., 2016a; Medeiros et al., 2019) allowed novel insights into the role of drought
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    tolerance traits in shaping drought response. The finding that PLA_{dry} and \pi_{tlp} can be useful for predicting
482
    drought responses of tree growth (Fig. 4; Table 1) is both novel and consistent with previous studies
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    linking these traits to habitat and drought tolerance. Previous studies have demonstrated that \pi_{tlp} and
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    PLA_{dry} are physiologically meaningful traits linked to species distribution along moisture gradients
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    (Maréchaux et al., 2015; Fletcher et al., 2018; Medeiros et al., 2019; Simeone et al., 2019; Rosas et al., 2019;
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    Zhu et al., 2018), and our findings indicate that these traits also influence drought responses. Furthermore,
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    the observed linkage of \pi_{tlp} to Rt in this forest aligns with observations in the Amazon that \pi_{tlp} is higher
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    in drought-intolerant than drought-tolerant plant functional type. Further, it adds support to the idea that
    this trait is useful for categorizing and representing species' drought responses in models (Powell et al.,
490
    2017). Because both PLA_{dry} and \pi_{tlp} can be measured relatively easily (Bartlett et al., 2012; Scoffoni
491
    et al., 2014), they hold promise for predicting drought growth responses across diverse forests. The
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it is feasible to study drought responses for all dominant species in most boreal and temperate forests (e.g., this study), this becomes difficult to impossible for species that do not form annual rings, and for diverse 495 tropical forests. Although progress is being made for the tropics (Schöngart et al., 2017), a full linkage of 496 drought tolerance traits to drought responses would be invaluable for forecasting how little-known species and whole forests will respond to future droughts (Christoffersen et al., 2016; Powell et al., 2017). 498 As climate change drives increasing drought in many of the world's forests (Trenberth et al., 2014; 499 Intergovernmental Panel on Climate Change, 2015), the fate of forests and their climate feedbacks will be 500 shaped by the biophysical and physiological drivers observed here. Our results show that taller, more exposed trees and species with less drought-tolerant leaf traits will be most affected in terms of growth 502 during the drought year. Survival is linked to resistance and resilience [CHECK REFS] (DeSoto et al., 503 2020; Gessler et al., 2020), implying it may be influenced by the same factors. Indeed, while the influence of PLA_{dry} and π_{tlp} on drought survival remains to be tested, taller trees have lower survival (Bennett 505 et al., 2015; Stovall et al., 2019). As climate change-driven droughts affect forests worldwide, there is likely 506 to be a shift from mature forests with tall, buffering trees to forests with a shorter overall stature 507 (McDowell et al., 2020). At this point, species whose drought tolerance relies in part on existence within a 508 buffered microenvironment (e.g., F. grandifolia) could in turn become more susceptible. Here, the relative 509 importance of tree height per se versus crown exposure becomes crucial, shaping whether the dominant 510 trees of shorter canopies are significantly more drought tolerant because of their shorter stature, or whether 511 high exposure makes them as vulnerable as the taller trees of the former canopy. Studies disentangling the 512 influence of height and exposure on drought tolerance will be critical to answering this question. 513 Ultimately, distributions of tree heights and drought tolerance traits across broad moisture gradients 514 suggest that forests exposed to more drought will shift towards shorter stature and be dominated by 515 species with more drought-tolerant traits (Liu et al., 2019; Bartlett et al., 2016a; Zhu et al., 2018). Our study helps to elucidate the mechanisms behind these patterns, opening the door for more accurate 517 forecasting of forest responses to future drought. 518

importance of predicting drought responses from species traits increases with tree species diversity; whereas

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493

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530 Author Contribution

- KAT, IM, and AJT designed the research. Tree-ring chronologies were developed by RH under guidance of
- AJT and NP. Trait data were collected by IM, JZ under guidance of NK and LS. Other plot data were
- collected by IM, AS, EGA, and NB under guidance of EGA and WM. Data analyses were performed by IM
- under guidance of KAT and VH. KAT and IM interpreted the results. IM and KAT wrote the first draft of
- manuscript, and all authors contributed to revisions.

536 Supplementary Information

- 557 NEED TO UPDATE !!** Table S1. Monthly Palmer Drought Severity Index (PDSI), and its rank
- among all years between 1950 and 2009 (driest=1), for focal droughts.
- Table S2. Species-specific bark thickness regression equations.
- Table S3. Species-specific height regression equations.
- Table S4. Individual tests of species traits as drivers of drought resistance, where Rt is used as the
- response variable.
- Table S5. Individual tests of species traits as drivers of drought resistance, where Rt_{ARIMA} is used as the
- response variable.
- Table S6. Summary of top full models for each drought instance, where Rt is used as the response variable.
- Table S7. Summary of top models for each drought instance, where Rt_{ARIMA} is used as the response
- variable.
- Figure S1. Time series of Palmer Drought Severity Index (PDSI) for the 2.5 years prior to each focal
- 549 drought
- 550 Figure S2. Map of ForestGEO plot showing topographic wetness index and location of cored trees
- Figure S3. Distribution of reconstructed tree heights across drought years.
- Figure S4. Distribution of independent variables by species.
- Figure S5. Comparison of Rt and Rt_{ARIMA} results, with residuals, for each drought scenario
- Figure S6. Visualization of best model, with data, for all droughts combined.
- 555 Appendix S1. Further Package Citations

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